Tecnológico de Costa Rica

High Performance Embedded Systems

Project 3

Optimizing Traditional and Deep Learning Algorithms for Autonomous Driving

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SGBM Algorithm Profiling

Given that the result obtained when attempting to profile with valgrind the provided implementation of the SGBM algorithm is that it takes a lot of time just to start the the Graphical User Interface (GUI) in the provided implementation, it was required to port the implementation to a console-based application. After implementing this, it was possible to obtain good results in the profiling.
Part of the most important results are presented in the Illustration 1.

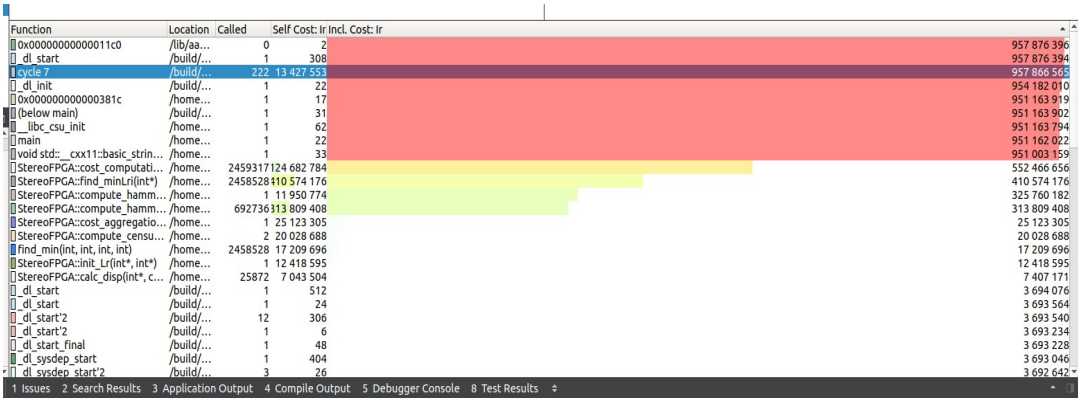

Illustration 1: Results after profiling the SGBM algorithm.

Given this, it was considered that the three most computation-expensive functions are:

- cost_computation.
- find_minLri.
- compute_hamming (really close with compute_hamming_distance).

Memory optimization

Once the algorithm was profiled, the next step was to optimize the amount of memory based on some assumptions in the input parameters.

Assuming that the input parameters are restricted to:

- BlockSize: 5 pixels
- P2 : 128
- NumDir: 4
- $C_{máx}$ (p, d): Results from Equation 3

The following results are obtained:

For equation 2:

$$\text{Census Transform Bit depth} = \text{Blocksize} * \text{Blocksize} - 1$$
$$= 5 * 5 - 1$$
$$= 24$$

For equation 3:

$$C_{máx}(p, d) = \text{Log}_2(\text{Census Transform Bit depth})$$
$$= \text{Log}_2(24)$$
$$= 4.58$$

For equation 4:

$$L \leq C_{máx}(p, d) + P2$$
$$L \leq 4.58 + 128$$
$$L \leq 132.58$$

For equation 6:

$$S \leq \text{Num. Dir} * (C_{máx}(p, d) + P2)$$
$$S \leq 4 * 132.58$$
$$S \leq 530.32$$